



# Analysis of the US-CERT DAC

Josh McNutt < imcnutt@cert.org >

FloCon: Netflow Analysis Workshop

July 21, 2004

CERT® Network Situational Awareness Group Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213-3890

The CERT Network Situational Awareness Group is part of the Software Engineering Institute. The Software Engineering Institute is sponsored by the U.S. Department of Defense.



| maintaining the data needed, and c<br>including suggestions for reducing  | lection of information is estimated to<br>ompleting and reviewing the collect<br>this burden, to Washington Headqu<br>uld be aware that notwithstanding ar<br>DMB control number. | ion of information. Send comments<br>arters Services, Directorate for Info | s regarding this burden estimate or or street | or any other aspect of th<br>, 1215 Jefferson Davis | nis collection of information,<br>Highway, Suite 1204, Arlington |  |
|---|---|--|---|---|--|--|
| 1. REPORT DATE 21 JUL 2004  |   | 2. REPORT TYPE   |   | 3. DATES COVERED <b>00-00-2004 to 00-00-2004</b>    |  |  |
| 4. TITLE AND SUBTITLE   |   | 5a. CONTRACT NUMBER  |   |   |  |  |
| Analysis of the US-CERT DAC   |   |  |   | 5b. GRANT NUMBER                                    |  |  |
|   |   |  |   | 5c. PROGRAM ELEMENT NUMBER                          |  |  |
| 6. AUTHOR(S)  |   |  |   | 5d. PROJECT NUMBER                                  |  |  |
|   |   |  |   | 5e. TASK NUMBER                                     |  |  |
|   |   |  |   | 5f. WORK UNIT NUMBER                                |  |  |
| 7. PERFORMING ORGANIZATION NAME(S) AND ADDRESS(ES)  Carnegie Mellon University,Software Engineering Institute,Pittsburgh,PA,15213 |   |  |   | 8. PERFORMING ORGANIZATION<br>REPORT NUMBER         |  |  |
| 9. SPONSORING/MONITORING AGENCY NAME(S) AND ADDRESS(ES)   |   |  |   | 10. SPONSOR/MONITOR'S ACRONYM(S)                    |  |  |
|   |   |  |   | 11. SPONSOR/MONITOR'S REPORT<br>NUMBER(S)           |  |  |
| 12. DISTRIBUTION/AVAIL <b>Approved for publ</b>   | LABILITY STATEMENT<br>ic release; distributi  | on unlimited   |   |   |  |  |
| 13. SUPPLEMENTARY NO presented at FloCo   | otes<br>on 2004, Crystal City   | y, VA, July 2004.  |   |   |  |  |
| 14. ABSTRACT  |   |  |   |   |  |  |
| 15. SUBJECT TERMS   |   |  |   |   |  |  |
| 16. SECURITY CLASSIFIC  | 17. LIMITATION OF   | 18. NUMBER   | 19a. NAME OF  |   |  |  |
| a. REPORT<br>unclassified   | b. ABSTRACT<br>unclassified   | c. THIS PAGE<br>unclassified   | Same as Report (SAR)  | OF PAGES 12   | RESPONSIBLE PERSON   |  |

**Report Documentation Page** 

Form Approved OMB No. 0704-0188





### Outline

- Data
- Graphical Displays
- Detecting Trends
- Anomaly Detection
- Roadmap





#### Data

#### Snort

- Signature-based alerts
- Pre-processor alerts

### Origin

- Multiple networks of varying size
- Volume
  - ~30-50 million alerts per month

### Ancillary Information

- Country code
- Netblock





# IDS Data: challenges

- No new attacks
  - Only matches known signatures
- Lack of context
  - Don't know what we are not seeing
- Non-standardized signature rule sets
  - No administrative control
- Missing Data
  - Uncertainty: Sensor failure vs. no intrusion attempts

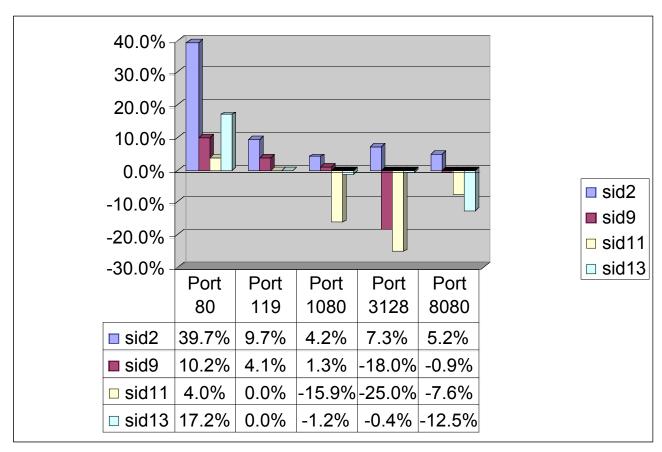




## TCP Destination Port Changes

Comparison of port activity across organizations shows monthly

trends.

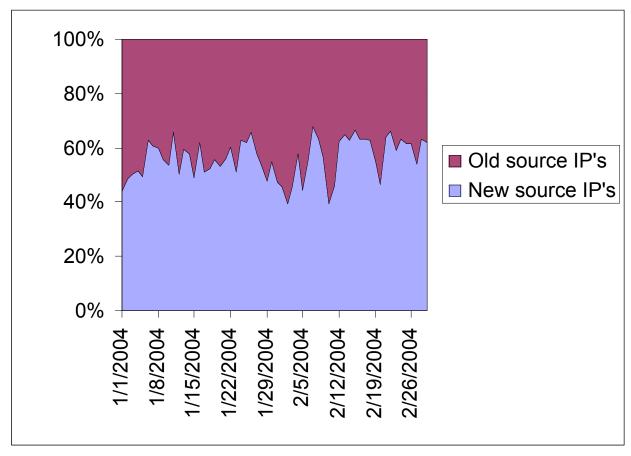






#### Share of New Source IP Addresses

Share of new daily source IP addresses stays fairly consistent.



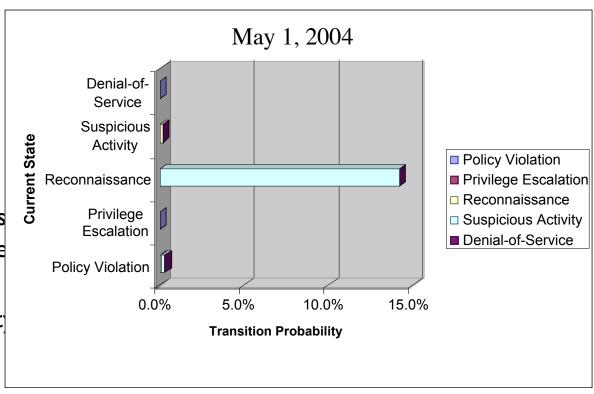




# Signature Class Transition

Transition probabilities highlight sequential patterns in data.

- Current State
  - Source IP records alert on Destination IP
- Transition probability
  - Percent chance for next class of alert recorded
- Most source/dest combos involve only one signature class
- Small transition probabilit for
  - Privilege Escalation

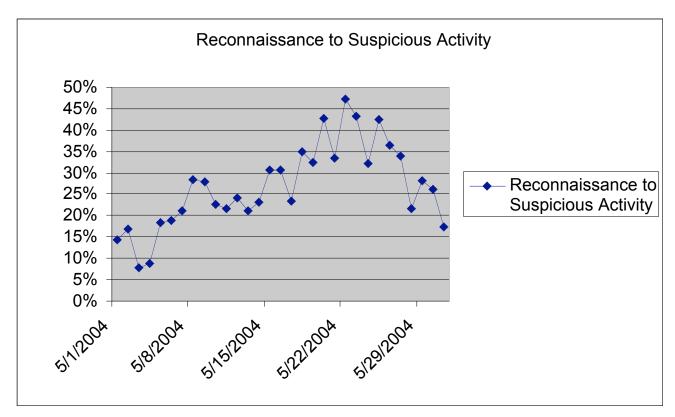






## Daily Transition Probabilities

Transition probabilities can be monitored over time to identify consistent sequences.



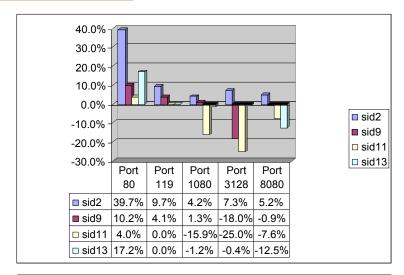


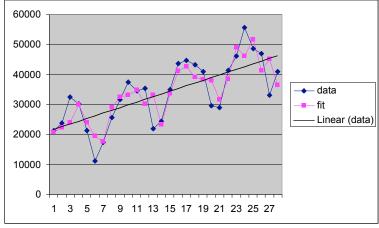


#### Trend Detection

- Current month vs. previous month
  - Across organizations
  - % changes

- Time Series
  - Fit trend line
    - Arbitrary time period
  - Seasonal Components
  - Regression with ARMA errors



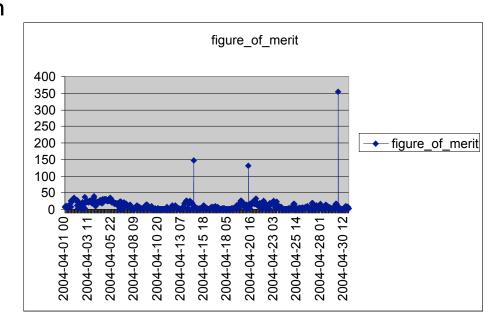






# **Anomaly Detection**

- Goal: Identify data points which deviate from overall pattern of data
- Our current implementation (Figure of Merit)
  - Evaluate hours
  - Record # alerts, # source IP addresses, # destination IP addresses, # signatures
- For each hour, we want measure of how deviant it was.

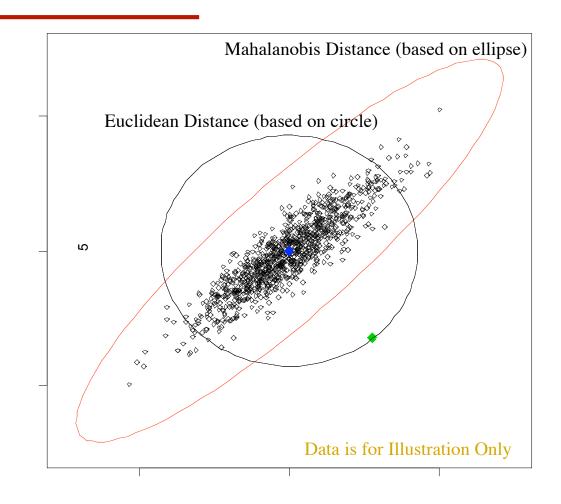






### Mahalanobis distance: 2D case

- Compute distance metric between each hour and the average hour
- When measuring Euclidean (Mahalanobis) Distance, all points along circle (ellipse) are same distance from the center
  - Points on larger circle/ellipse are greater distance from center
- Shape of the ellipse
  - Function of correlation between variables
- Generalizes to n dimensions (Ellipsoid)







## Analysis Roadmap

- Incorporate flow data
- Automating trend detection
  - Time series analysis
- Clustering
  - Group sources by similar activity patterns
    - Temporal correlation
    - Targeting similarities
    - Signature usage
  - Look for evidence of possible coordination